



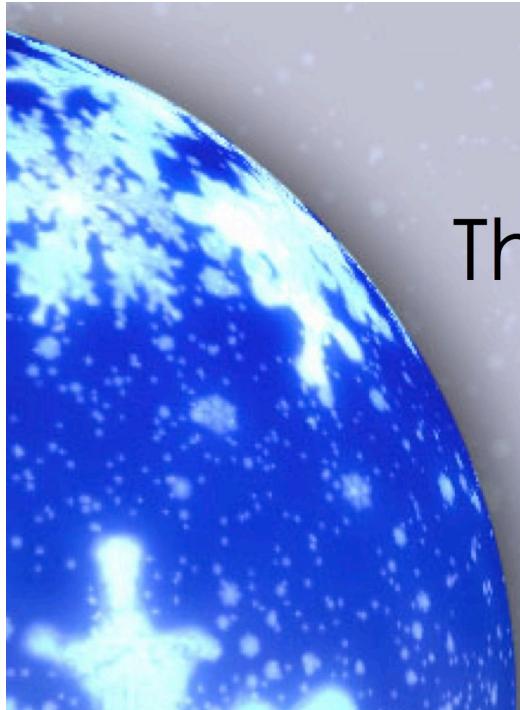
# A New Approach to Photometry for Mixed Resolution Data Sets

Anthony Gonzalez  
Lexi Moustakas



# PyGFIT: A New Approach to Photometry for Mixed Resolution Data Sets

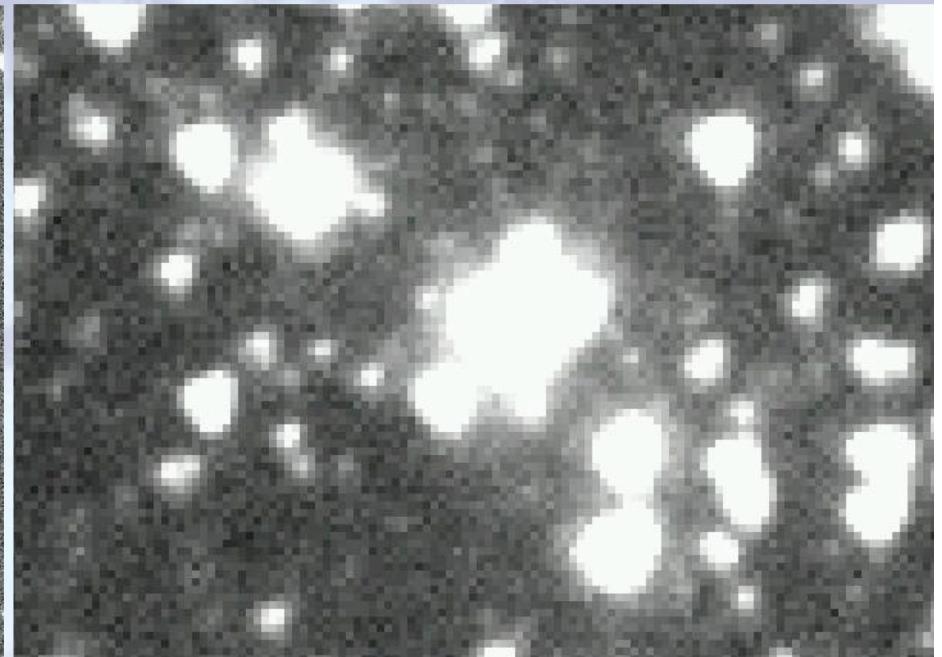
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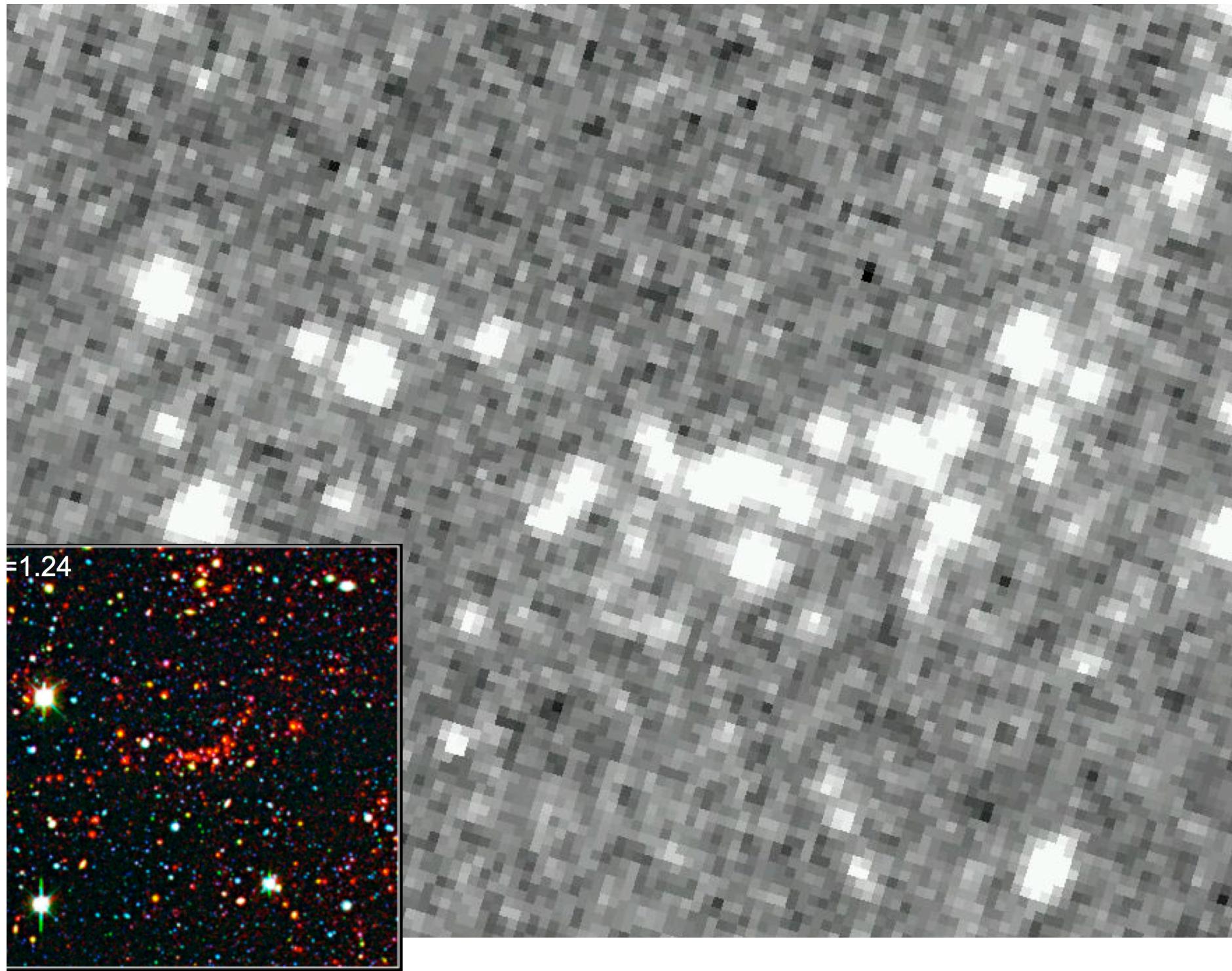


# The Question

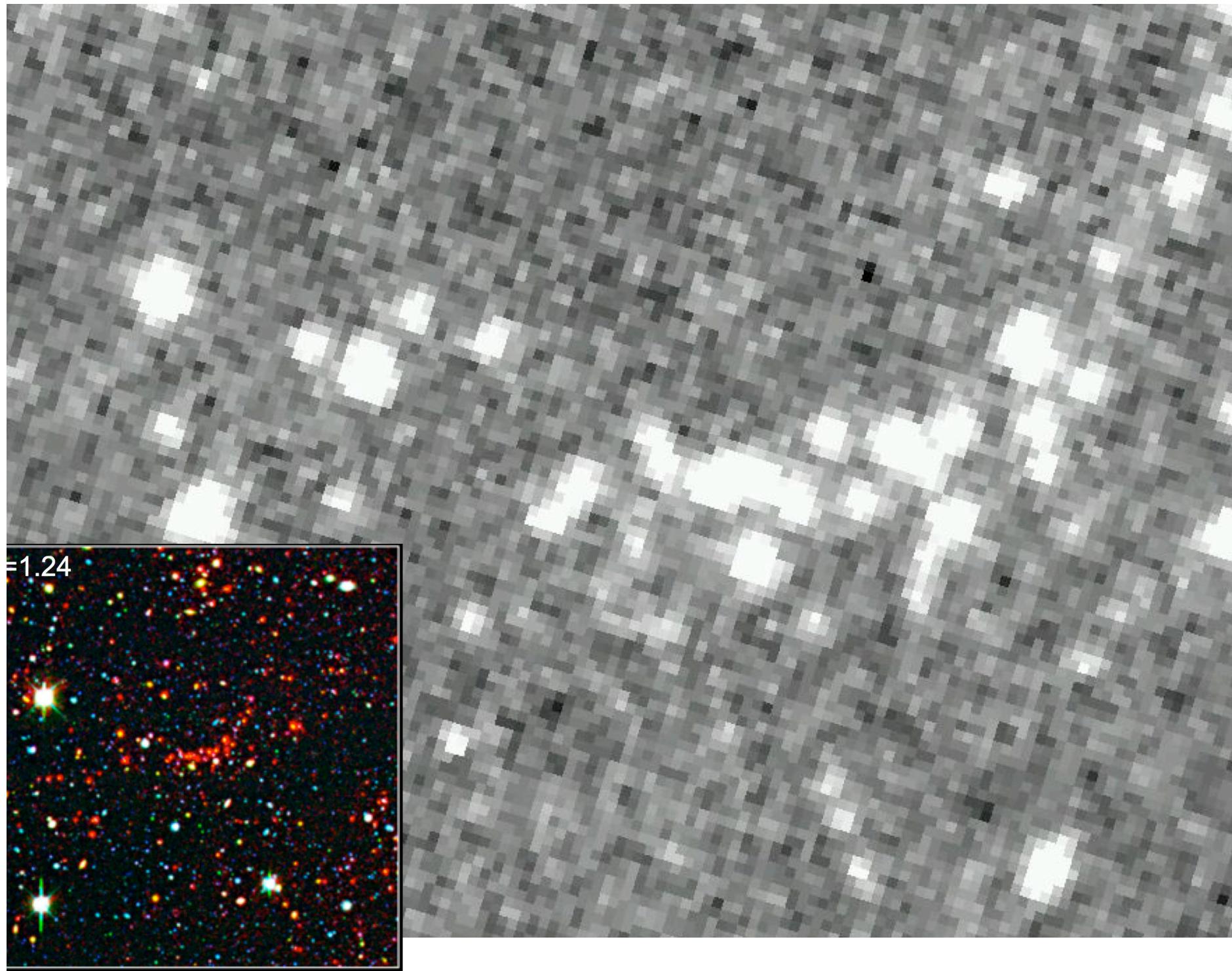


- ⌘ How to best measure photometry in different resolution images?

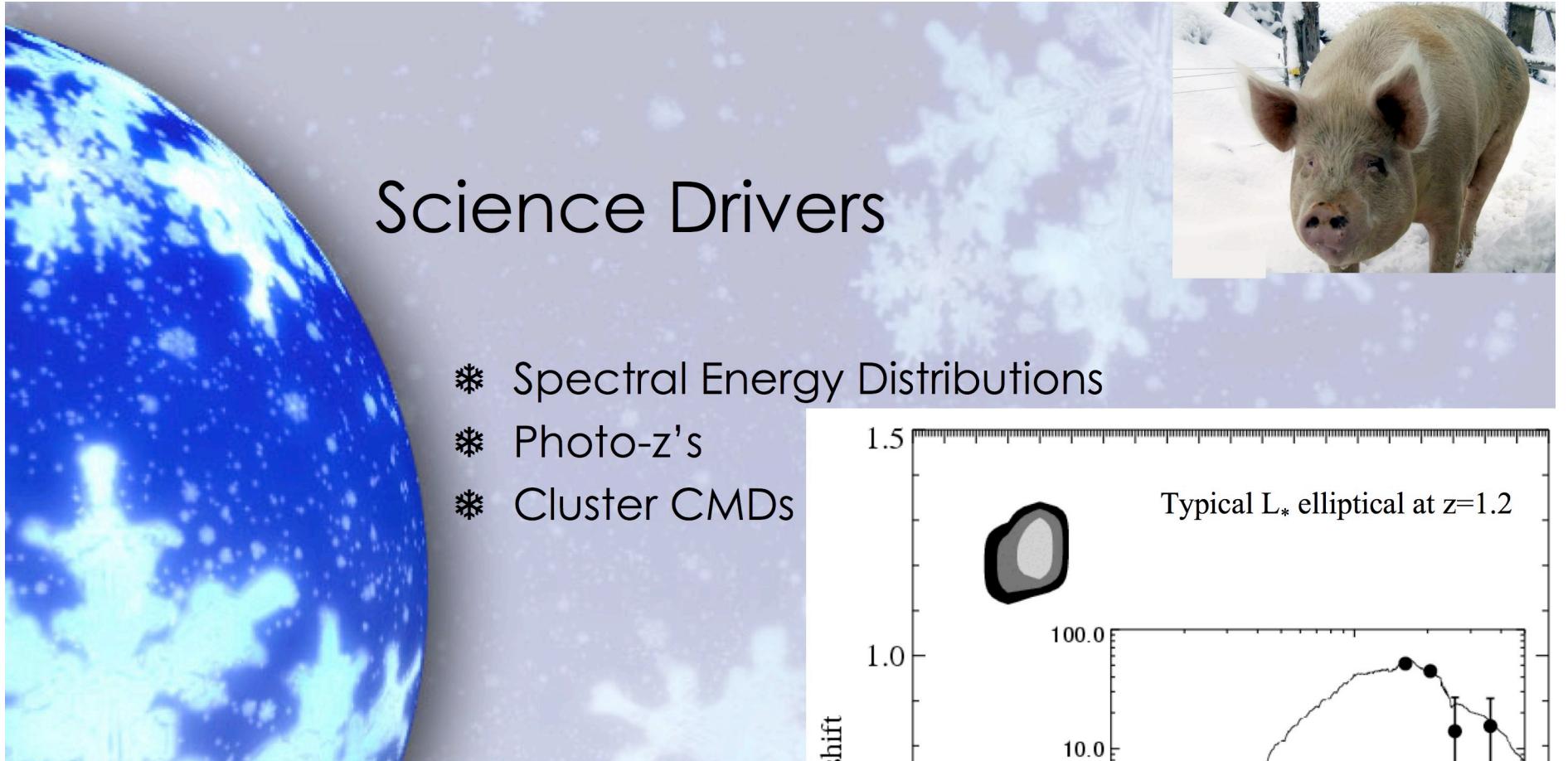




=1.24

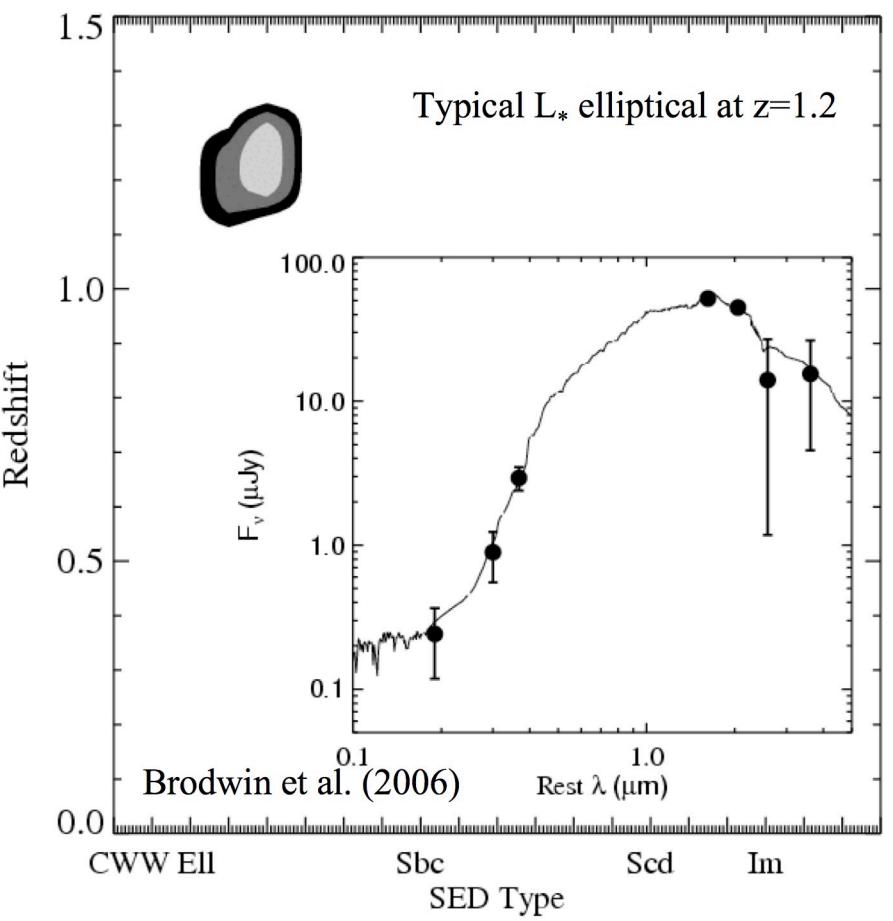
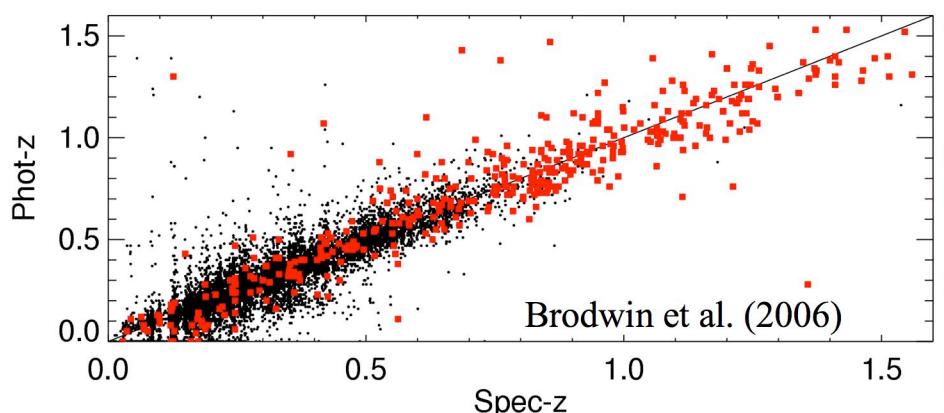


=1.24



# Science Drivers

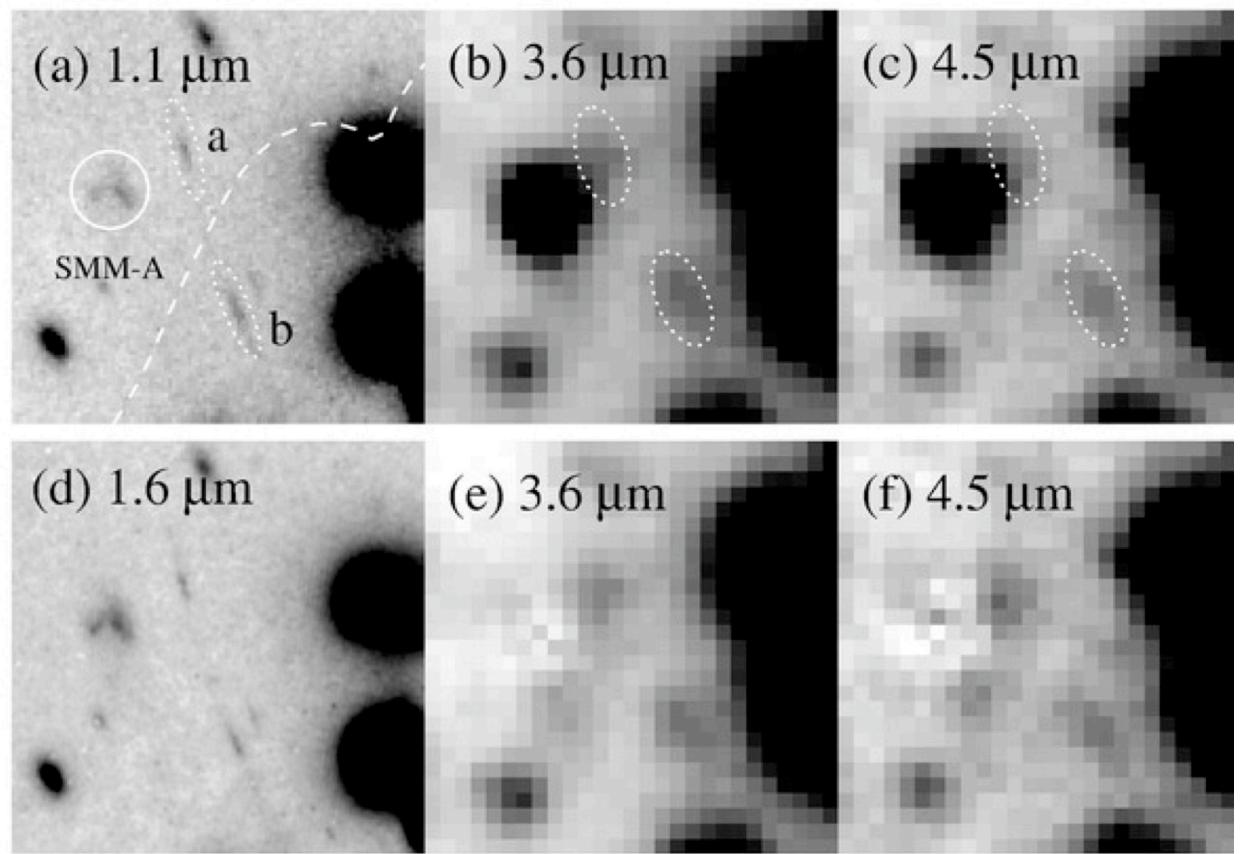
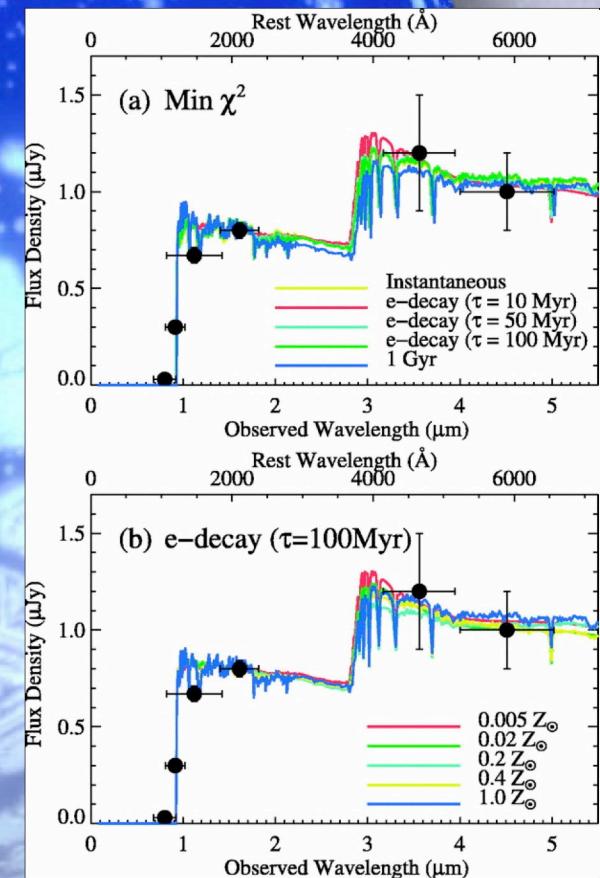
- ✿ Spectral Energy Distributions
- ✿ Photo-z's
- ✿ Cluster CMDs





# Science Drivers

⌘ Removal of bright foregrounds



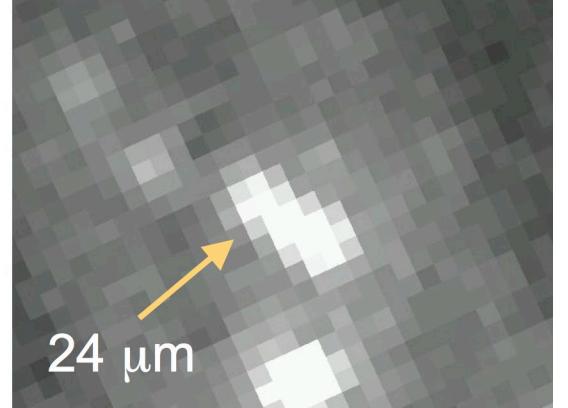
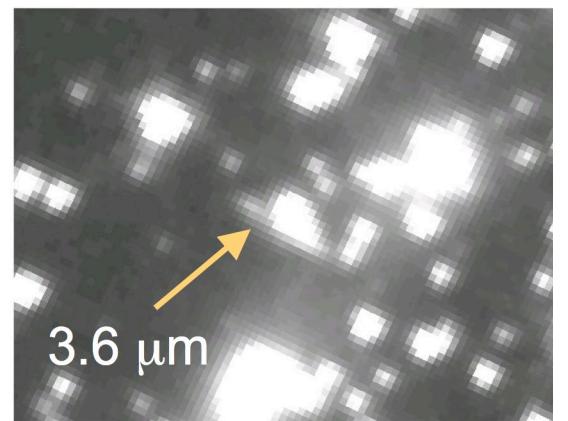
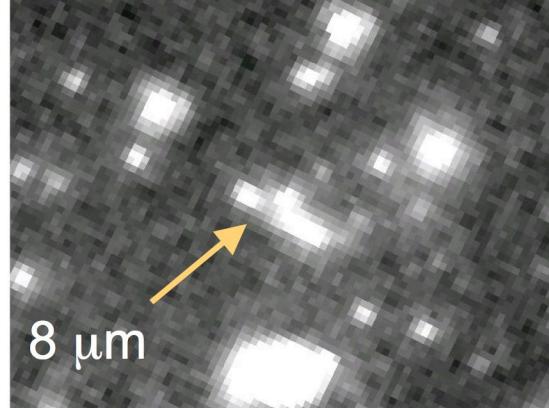
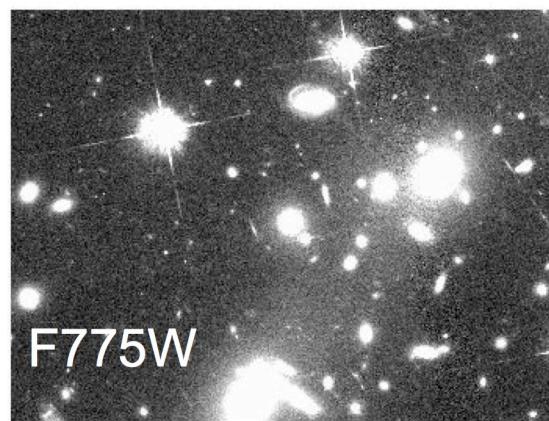
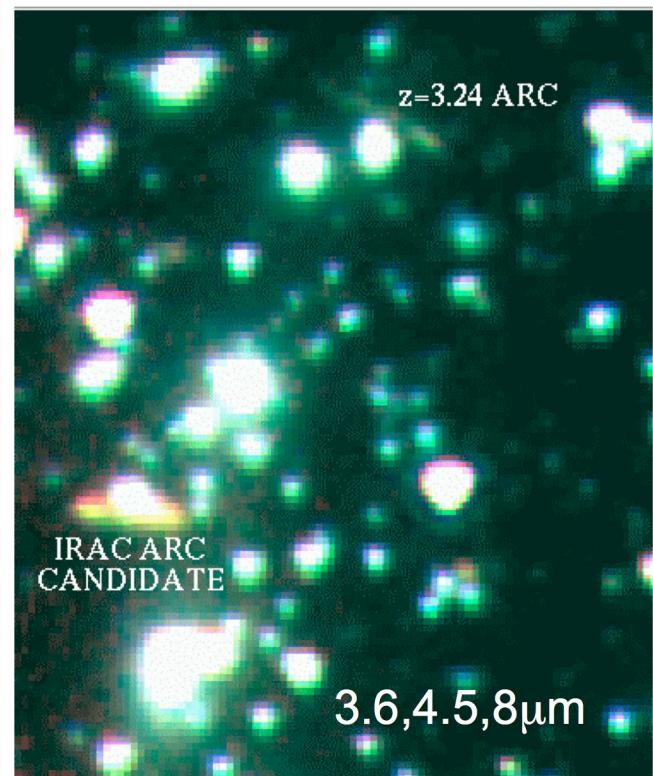
Egami et al. (2005)



# Science Drivers



⌘ Removal of bright foregrounds

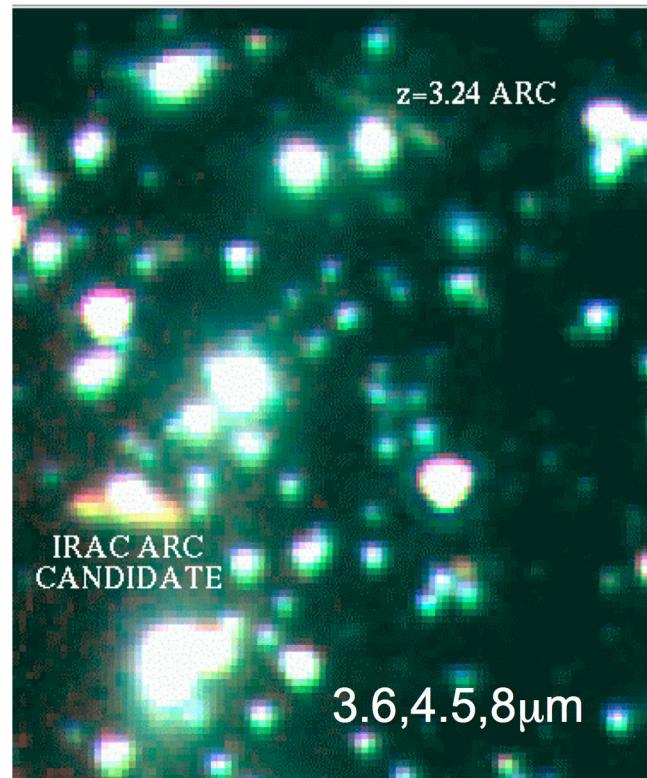


Gonzalez et al. (2008), in prep

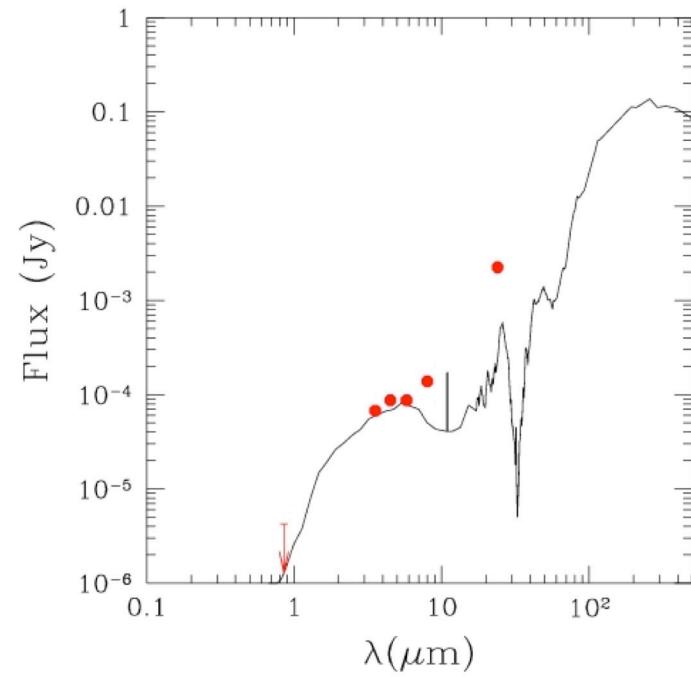
# Science Drivers



⌘ Removal of bright foregrounds



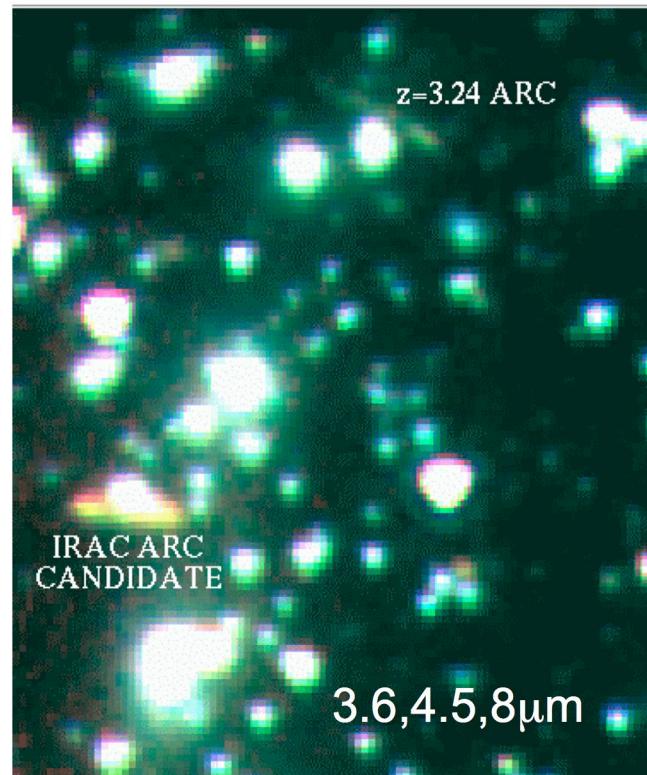
Gonzalez et al. (2008), in prep



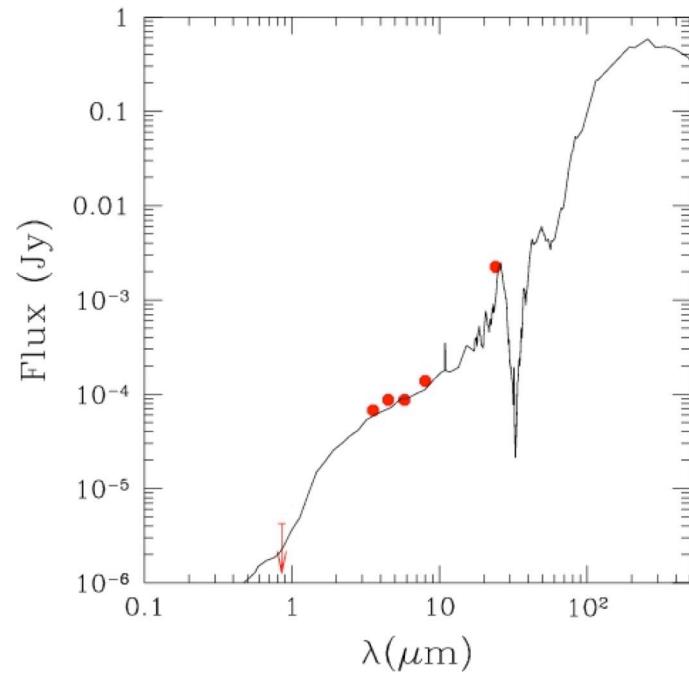
# Science Drivers



⌘ Removal of bright foregrounds



Gonzalez et al. (2008), in prep



$\sim 5x$  high TIR component than Arp 220

# Non-parametric Codes



⌘ TFIT

⌘ Laidler et al 2007

⌘ ConvPhot

⌘ de Santis et al. 2007, Grazian et al. 2006

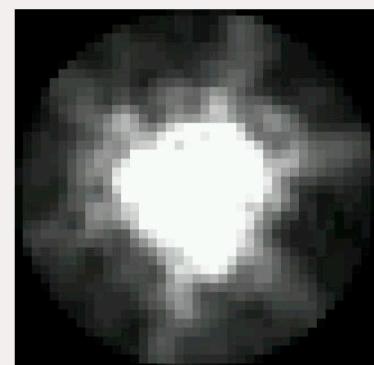
$$\sum a_i T_i * PTF \rightarrow Model$$



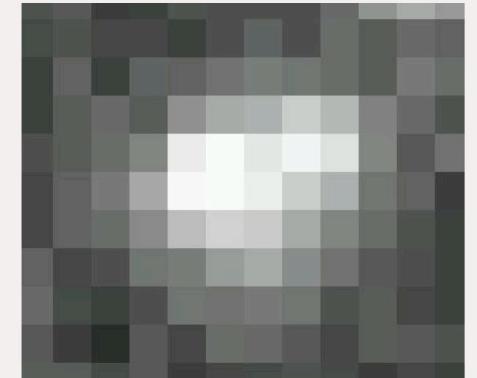
$T_i$



\*



$\chi^2$  minimization



*Must rebin to match resolution*

PSF Transfer Function (PTF)

Model

# Non-parametric Codes



Key Advantage: Direct Comparison of data sets, so minimal assumptions

\* ConvPhot

\* de Santis et al. 2007, Grazian et al. 2006

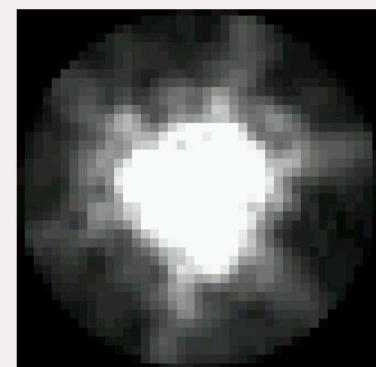
$$\sum a_i T_i * PTF \rightarrow Model$$



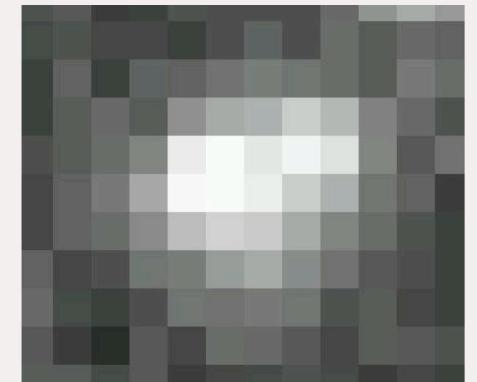
$T_i$



\*



$\chi^2$  minimization



*Must rebin to match resolution*

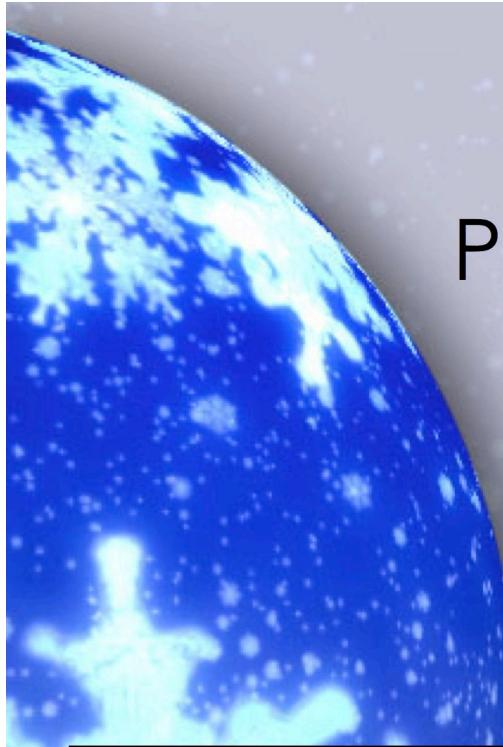
PSF Transfer Function (PTF)

Model

# General Considerations



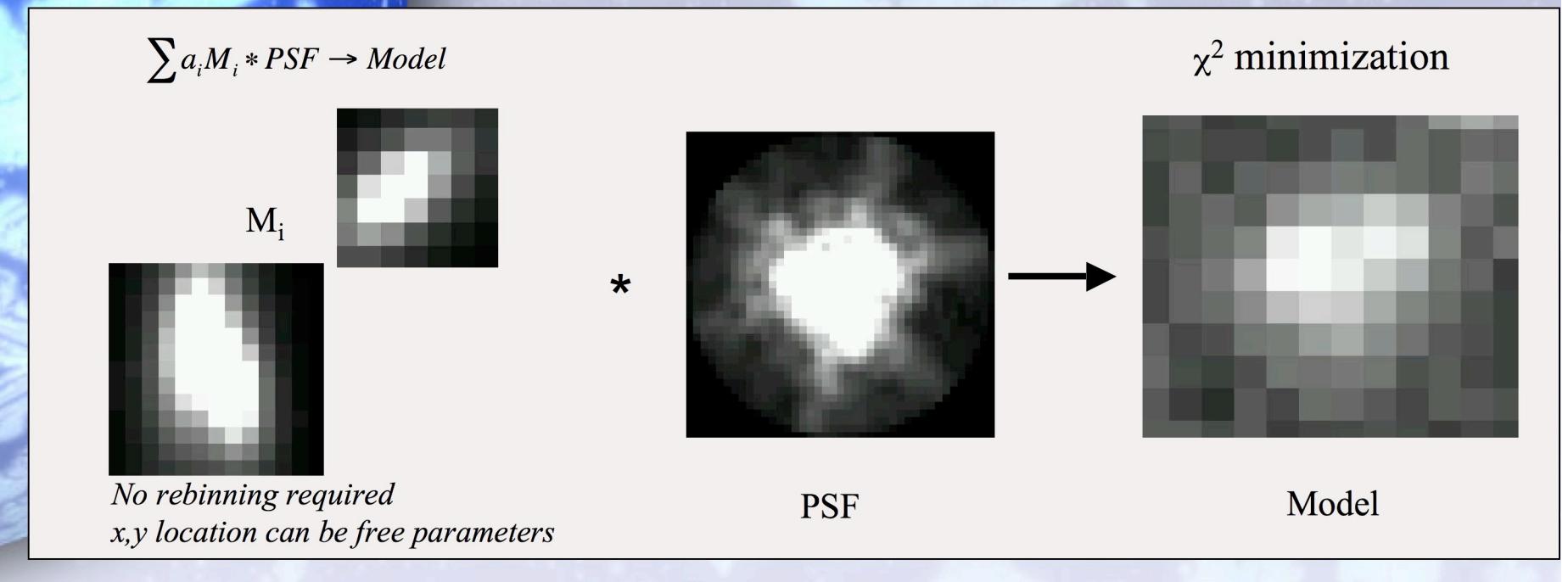
- ⌘ Limitations
  - ⌘ High-resolution data must be deep
  - ⌘ Objects must be well-separated in detection image
- ⌘ Assumptions and Potential Biases
  - ⌘ Morphological k-corrections
  - ⌘ Systematics due to PSF mismatch
- ⌘ Astrometric and alignment issues
  - ⌘ Pixelization/Resampling effects
  - ⌘ Jitter/Subpixel offsets between images
  - ⌘ Must set images to same orientation and rebin to appropriate resolution



# PyGFit Approach



- ✿ General concept similar to TFIT & ConvPhot
- ✿ Parametric using galaxy structural models
  - ✿ Works with standard quantitative morphology software (Galfit, GIM2D)
  - ✿ Can be used with any analytic model





# PyGFit Approach

- ⌘ Disadvantages
  - ⌘ Requires galaxy structural models
  - ⌘ (Reasonable) extrapolation beyond observed profiles
  - ⌘ Models will not include asymmetric components
    - ⌘ Irregular galaxies, tidal tails, ..
  - ⌘ Garbage in  $\Rightarrow$  Garbage out



# Advantages



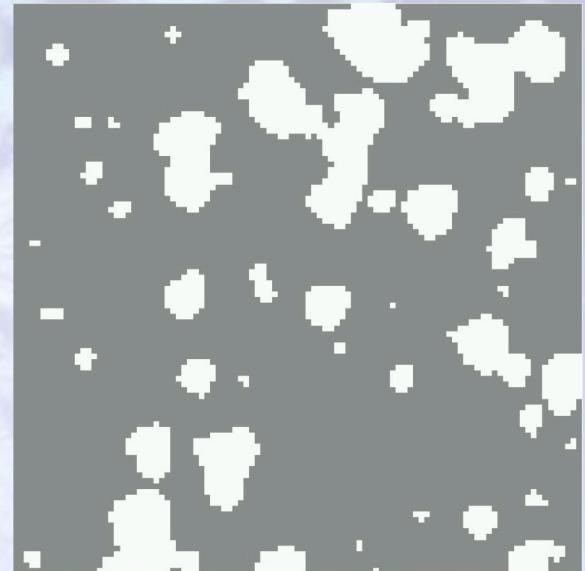
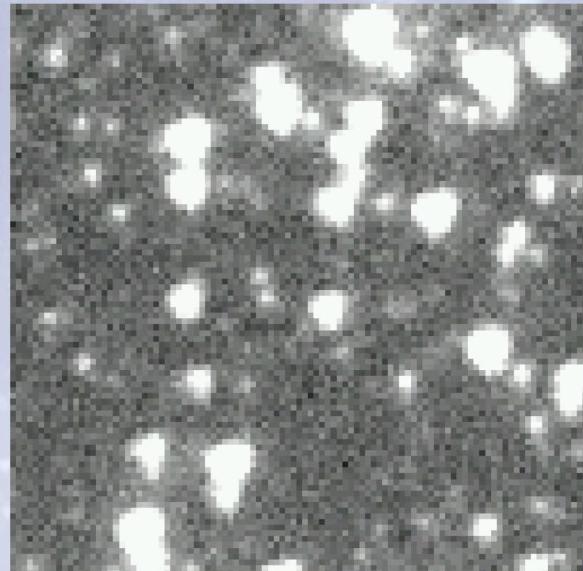
- ✿ **Easy to get running**
  - ✿ Need: image, rms image, PSF, structural catalog
- ✿ **Fast**
  - ✿ ~10x faster than ConvPhot (based on specs on ConvPhot web page)
  - ✿ Well-suited for large survey fields
- ✿ Alignment issues simple
- ✿ “Matched” photometry
  - ✿ Same model used for high and low resolution images
- ✿ Straightforward to simulate uncertainties
- ✿ Robust/Stable
  
- ✿ Works near saturated stars
- ✿ Can work at somewhat more crowded field levels
  - ✿ Sources can overlap in detection image
- ✿ Works on image boundaries
- ✿ If desired, can fit bulge, disk magnitudes independently



# Implementation

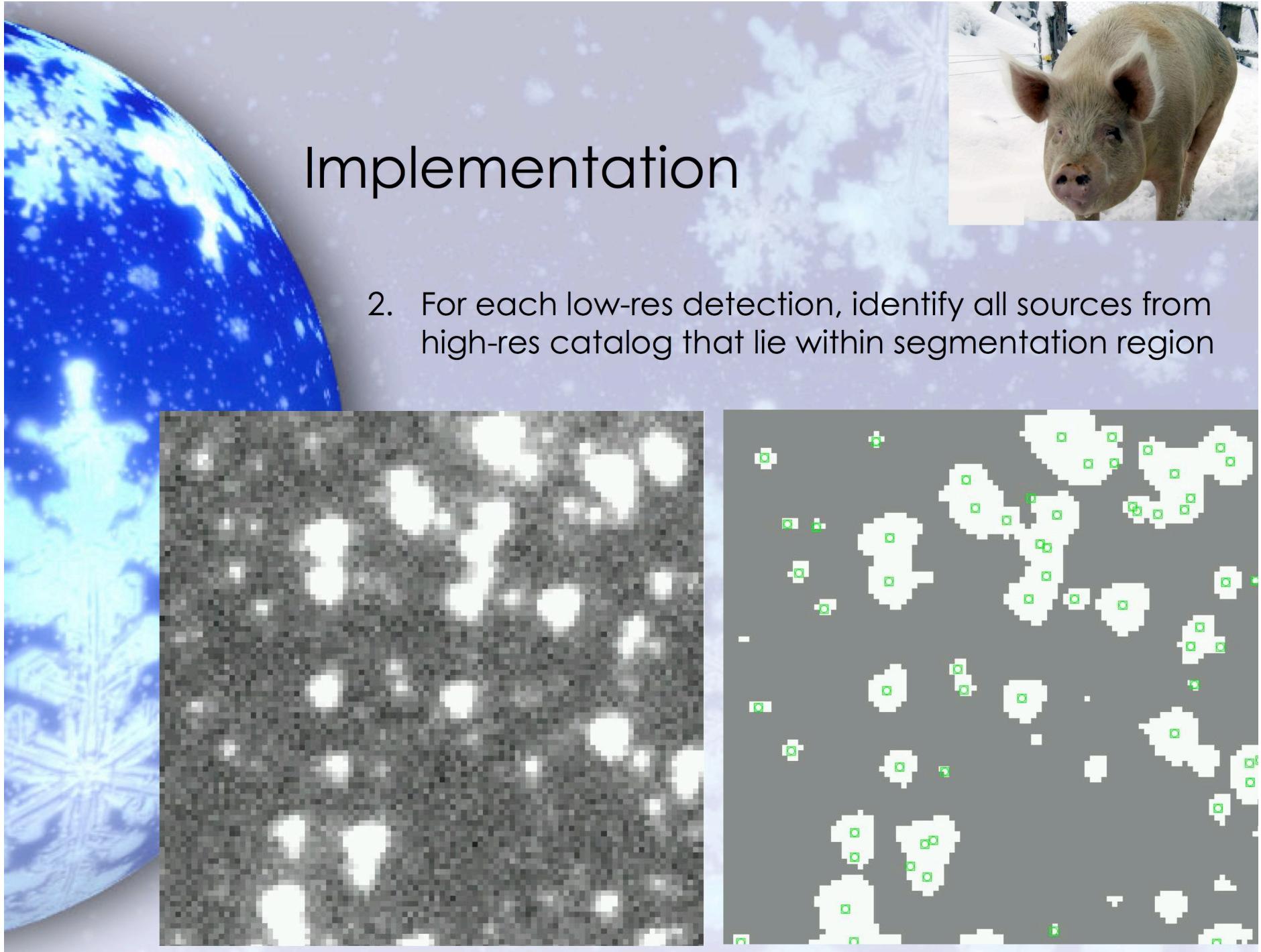


1. Background subtraction & object detection in low resolution image (SExtractor)



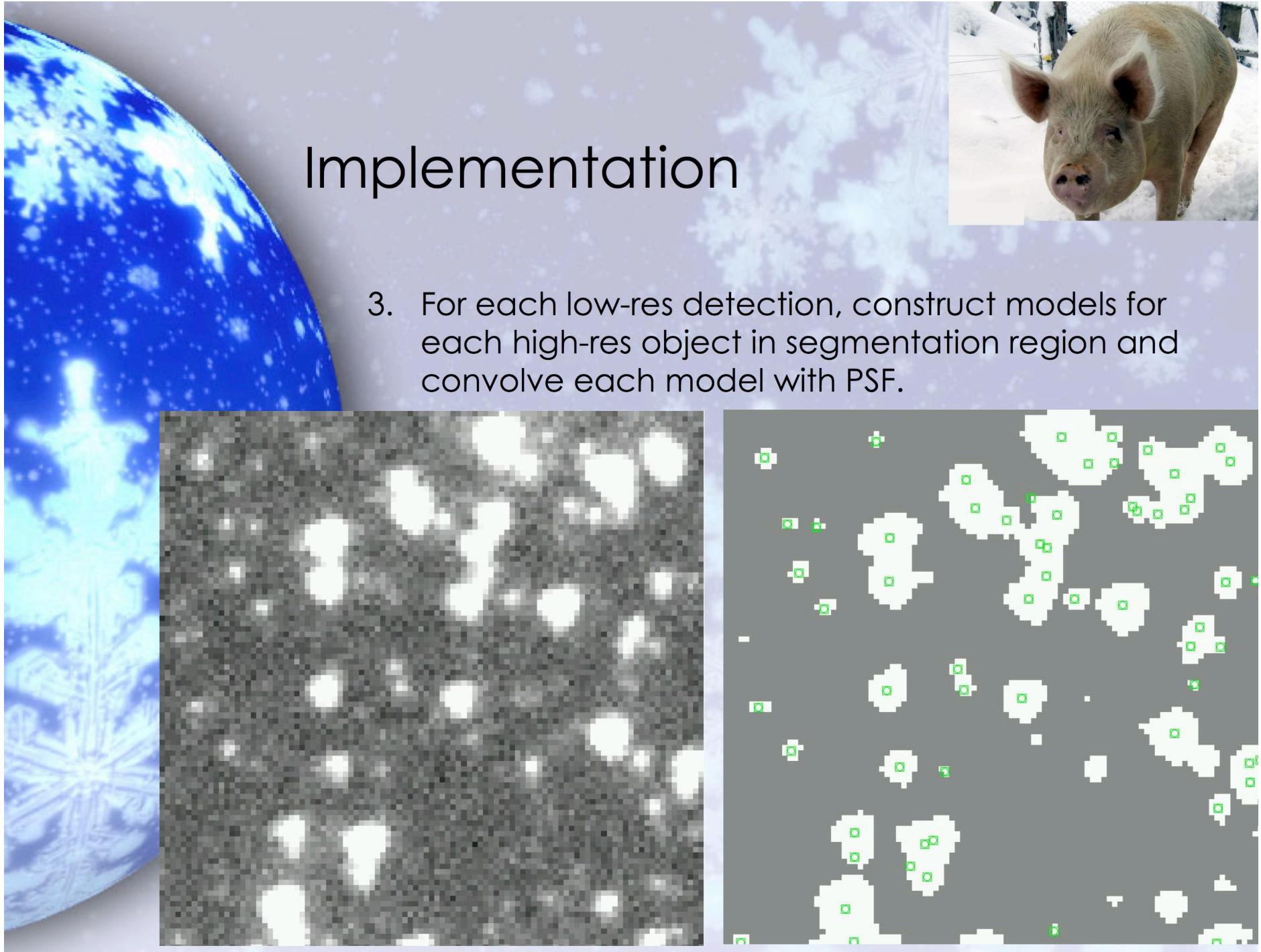
# Implementation

2. For each low-res detection, identify all sources from high-res catalog that lie within segmentation region



# Implementation

3. For each low-res detection, construct models for each high-res object in segmentation region and convolve each model with PSF.



# Implementation



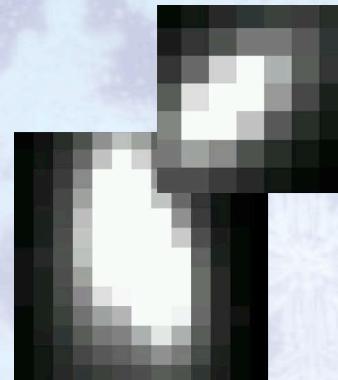
4. Run minimization
  1. Simultaneous fit to all objects in segmentation region
  2. Magnitude is free parameter
  3. x,y are constrained free parameters for each object; can also be linked if desired.



# Implementation



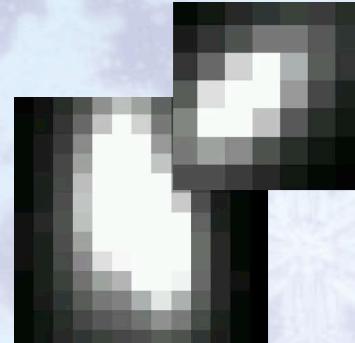
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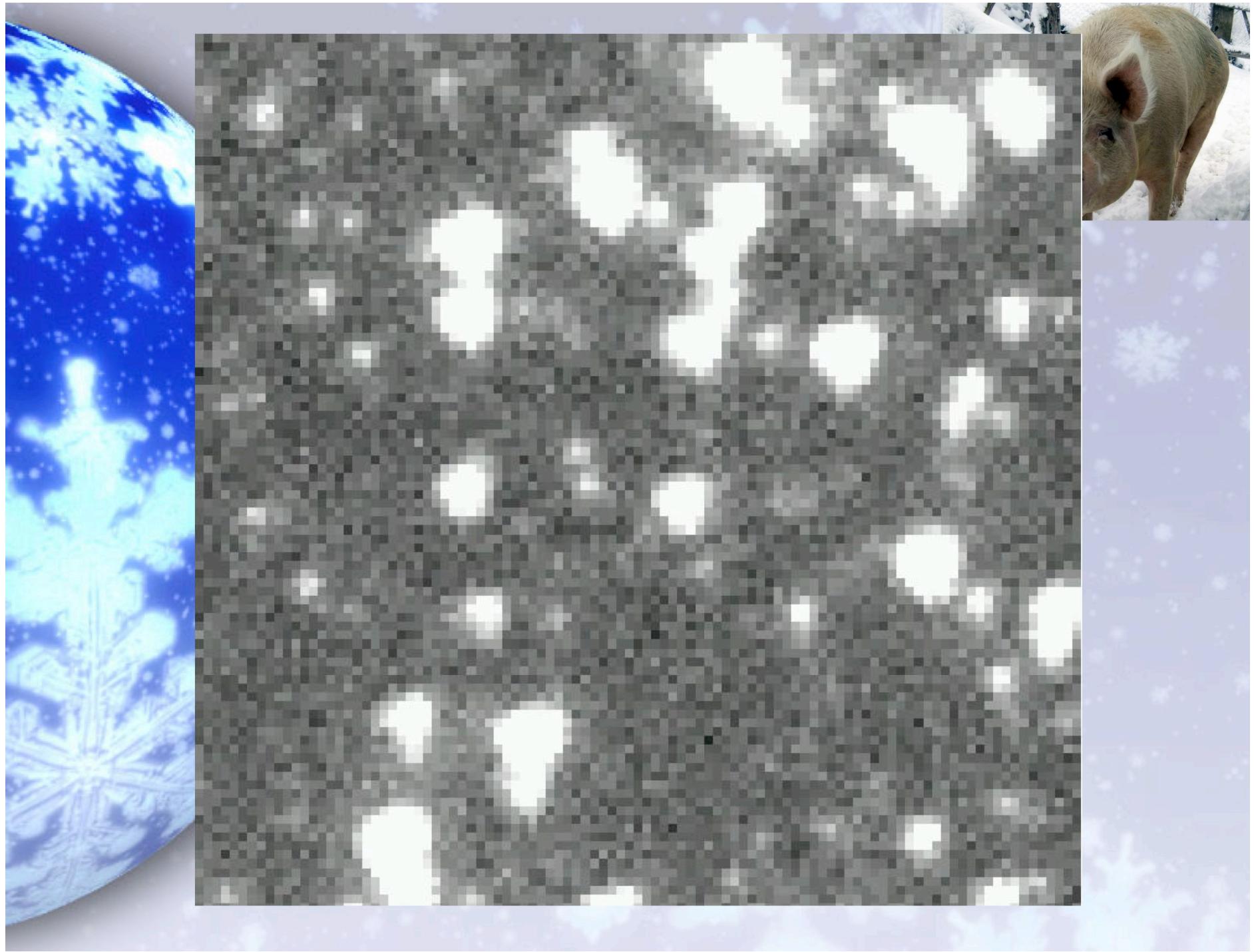


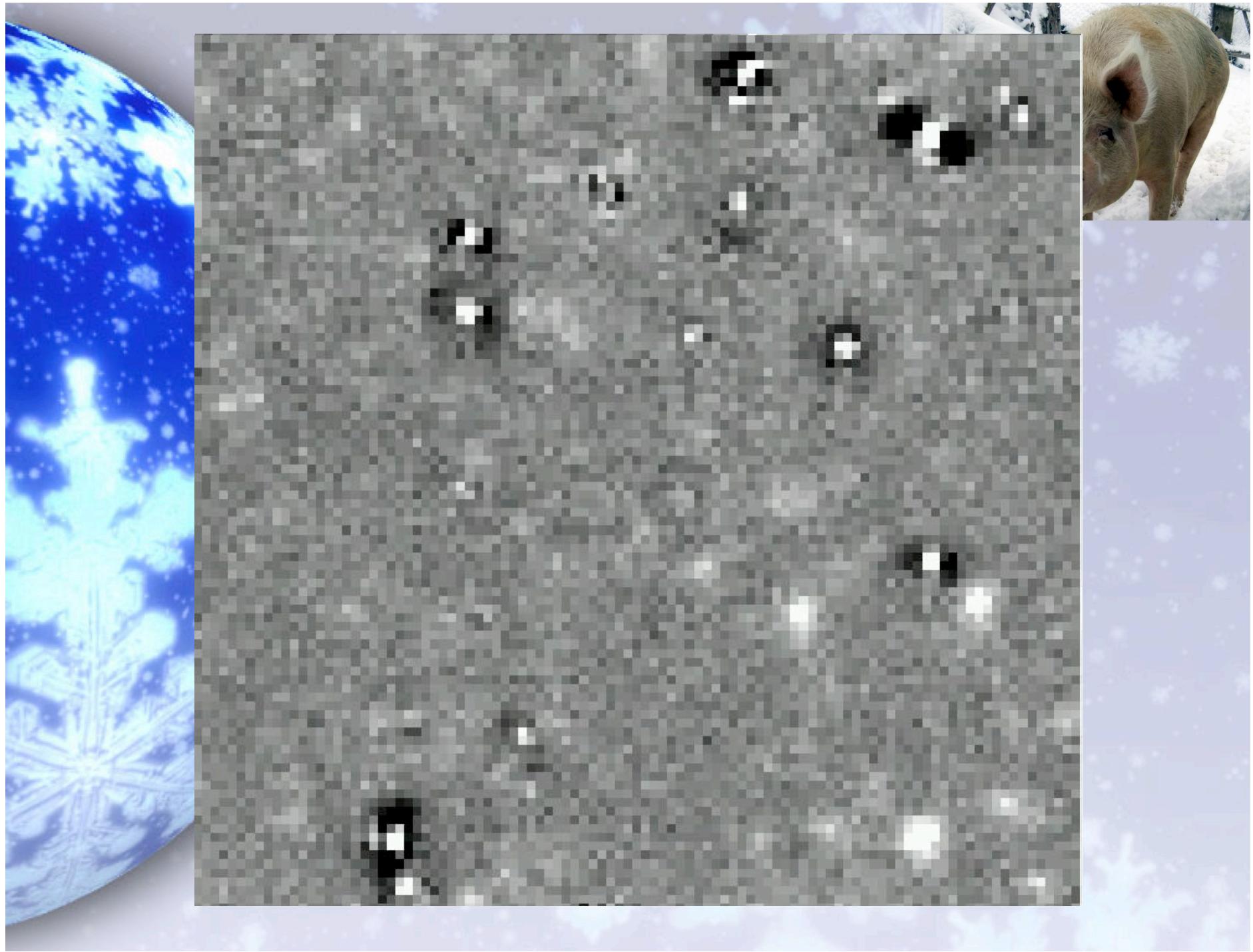
# Implementation

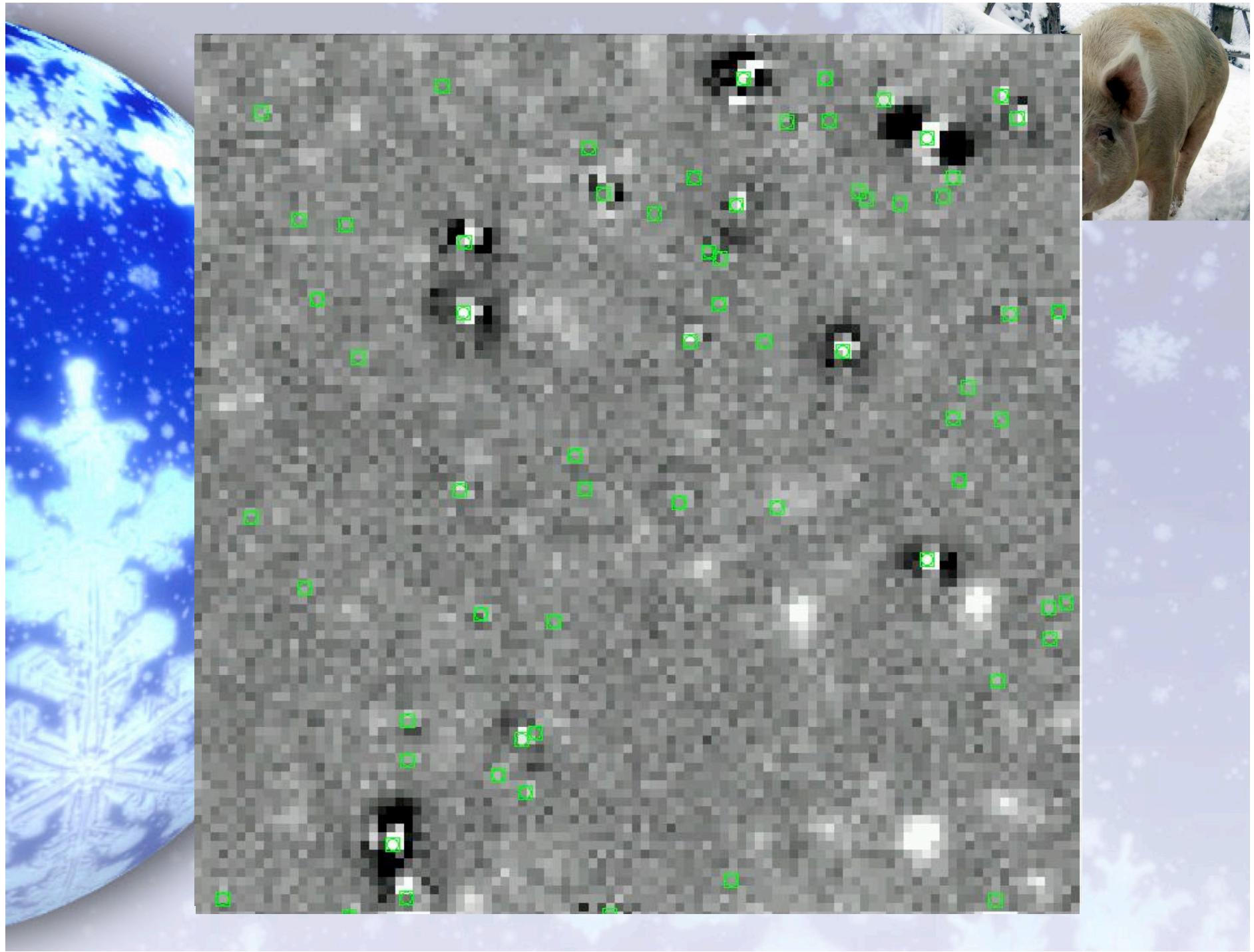


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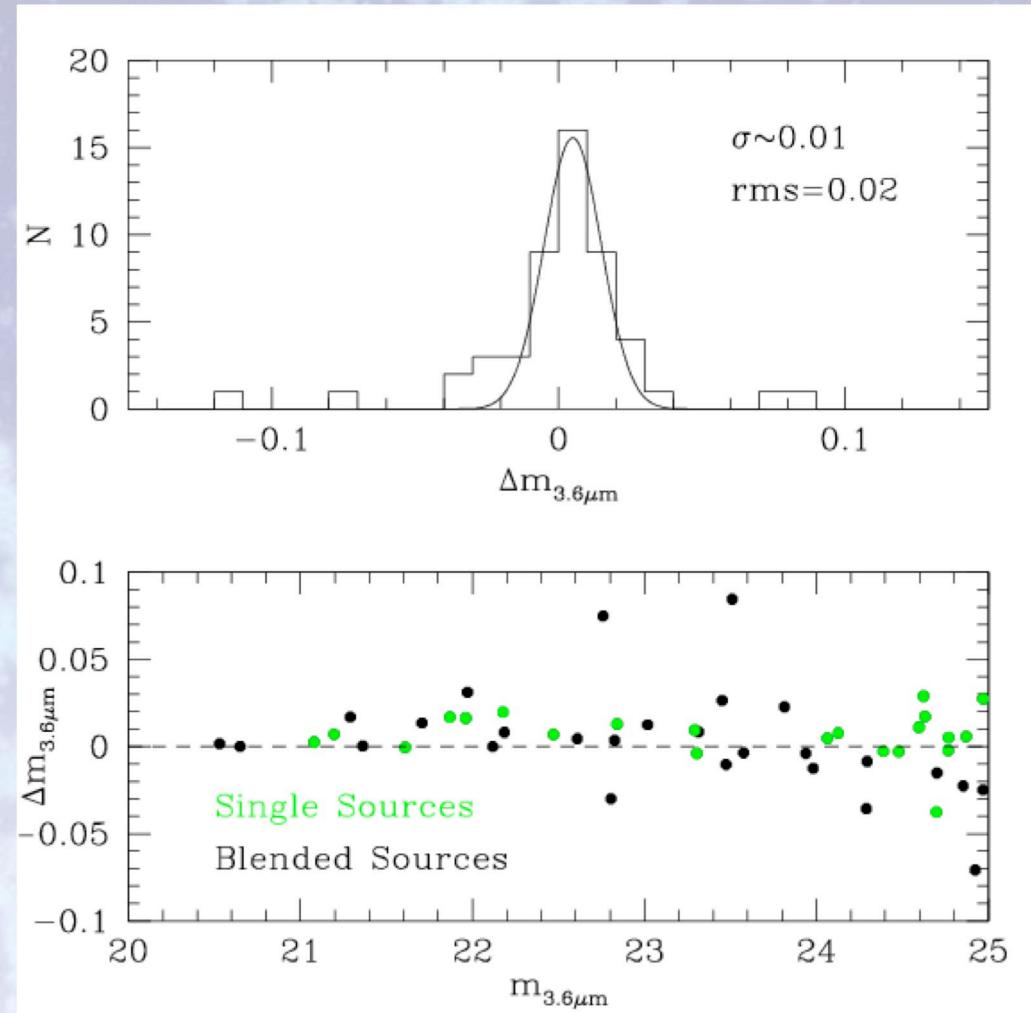
# Sensitivity to PSF



## GOODS IRAC Comparison

- Empirical vs SSC PSF
  - Consistent results for 94% of sources
  - RMS scatter of 0.02 mag between two input PSFS
- Use of random star in field as PSF:
  - RMS scatter of 0.06 mag
  - Consistent results for 91% of sources

*Results not strongly dependent upon input PSF.*



# Stability to X,Y Jitter

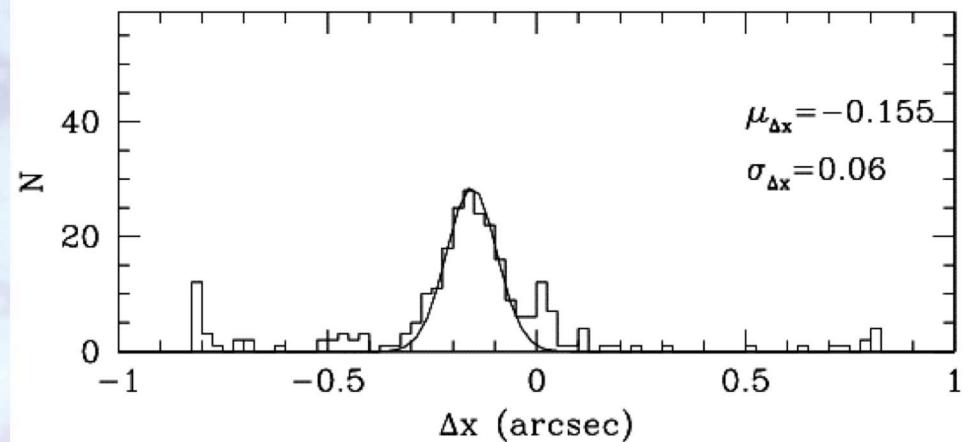
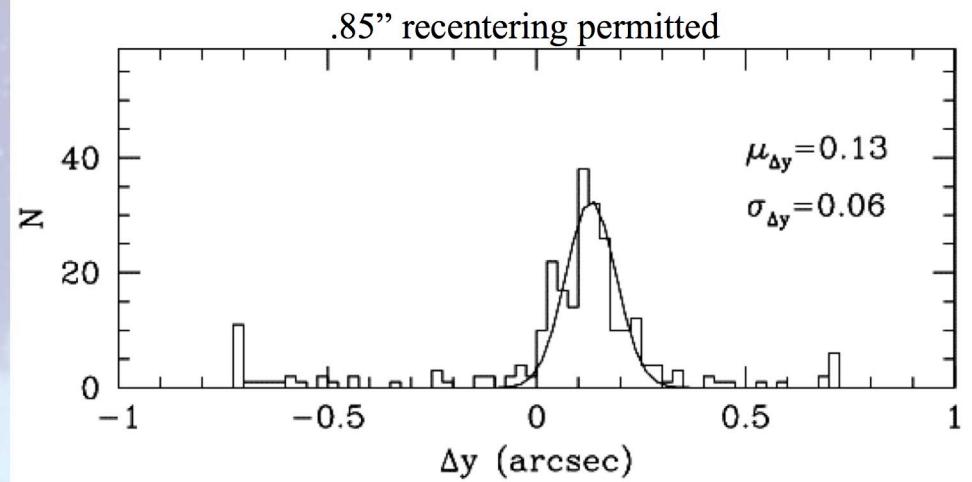


Allowing X,Y to vary independently for each ACS source:

Solutions robust to translational freedom

Typical astrometric scatter in solutions for GOODS IRAC+ACS data is  $0.06''$  along each axis

Few % drift to boundaries of permitted recentering.



# Initial Science?



- ⌘ Code currently in alpha testing
- ⌘ Upcoming
  - ⌘ GOODS Comparisons
  - ⌘ Bullet Cluster (see also M. Bradac's talk)
  - ⌘ IRAC Shallow Survey Galaxy Clusters at  $z=1-1.5$   
(see M. Brodwin's talk)